





SARS-CoV-2 Genomic Surveillance Reveals Little Spread From a Large University Campus to the Surrounding Community

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Background. Coronavirus disease 2019 (COVID-19) has had high incidence rates at institutions of higher education (IHE) in the United States, but the transmission dynamics in these settings are poorly understood. It remains unclear to what extent IHE-associated outbreaks have contributed to transmission in nearby communities.

Methods. We implemented high-density prospective genomic surveillance to investigate these dynamics at the University of Michigan and the surrounding community during the Fall 2020 semester (August 16–November 24). We sequenced complete severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) genomes from 1659 individuals, including 468 students, representing 20% of cases in students and 25% of total cases in Washtenaw County over the study interval.

Results. Phylogenetic analysis identified >200 introductions into the student population, most of which were not related to other student cases. There were 2 prolonged student transmission clusters, of 115 and 73 individuals, that spanned multiple on-campus residences. Remarkably, <5% of nonstudent genomes were descended from student clusters, and viral descendants of student cases were rare during a subsequent wave of infections in the community.

Conclusions. The largest outbreaks among students at the University of Michigan did not significantly contribute to the rise in community cases in Fall 2020. These results provide valuable insights into SARS-CoV-2 transmission dynamics at the regional level. **Keywords.** genomic epidemiology; infection prevention; SARS-CoV-2; transmission; university.

Institutions of higher education (IHE) have been associated with high incidence of coronavirus disease 2019 (COVID-19) in the United States [1–3]. Congregate settings, such as on-campus housing and off-campus social gatherings, have led to large outbreaks despite prevention efforts [4–8]. It is essential to gain a better understanding of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) transmission dynamics surrounding IHE to inform prevention strategies [9–11], especially as more contagious variants circulate. An important question is to what extent IHE-related outbreaks have contributed to transmission in the communities where IHE are geographically located. One study showed that counties with large

IHE that opened for in-person learning had higher incidence of COVID-19 compared with matched counties with remote-only learning [2]. However, it is possible that this reflects transmission into and among student populations without spread into surrounding communities. Some studies have analyzed case counts and anonymized behavioral and movement data to assess mixing between populations, but these approaches are limited in their ability to track specific transmission chains [12, 13]. Complementary approaches, such as contact tracing and genomic epidemiology, may be useful to more directly assess whether IHE-related outbreaks have spread into nearby communities.

Virus genome sequencing has been an important epidemiologic tool during the COVID-19 pandemic, enabling the characterization of transmission lineages and their connections across different populations [14–16]. Phylogenetic analysis from well-sampled populations can reveal the number of unique introductions of SARS-CoV-2, the growth and persistence of lineages, and the frequency of transmission crossover between groups. An important advantage of genomic surveillance is that it can rule out or establish transmission relatedness between groups that appear to be associated through

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analyses of case counts and onset dates. However, the power of genomic surveillance to answer these questions relies on dense, comprehensive sampling in the populations under investigation.

Recent studies suggest that spillover from outbreaks at IHE occurs infrequently, but there are relatively few that have used high-density genomic surveillance. A genome sequencing study described the transmission of SARS-CoV-2 from an outbreak among students in La Crosse County, Wisconsin, into skilled nursing facilities, resulting in 2 deaths [17], but other studies have noted limited transmission from students. A seroprevalence study in Pennsylvania indicated that the community incidence of COVID-19 remained low throughout a period of high incidence in students [18]. Few or no viral descendants of student clusters in Dane County, Wisconsin, and Washington State have been detected in genomic surveillance of their respective communities with sampling density of 1%-3% [4, 6]. One notable study from a university in the United Kingdom used genomic surveillance (8% of the community) and contact tracing to demonstrate separate viral lineages in university vs community cases [19]. Because these studies have varied greatly in the depth and breadth of community sampling, the extent of transmission between students and the community in other settings is not clear.

To address this question, we conducted prospective, high-density genomic SARS-CoV-2 surveillance during the Fall 2020 semester at the University of Michigan-Ann Arbor and the surrounding area. We sequenced complete genomes from 1659 individuals, including 468 students and 1191 nonstudents. We captured 20% of confirmed cases in University of Michigan-Ann Arbor students and roughly one-quarter of confirmed cases in Washtenaw County, where the university is located. We detected >200 independent transmission introductions into the student population, including 2 large lineages that persisted for several weeks. However, there was very little crossover from student lineages into the community. The largest student-associated lineages waned by mid-November 2020, when community incidence drastically increased. We conclude that large outbreaks among students at the University of Michigan-Ann Arbor did not significantly drive the increase in community COVID-19 incidence in Southeastern Michigan in November 2020.

METHODS

Research Ethics and Sample Sources

Use of residual SARS-CoV-2-positive specimens from Michigan Medicine laboratories and collection of student status and on-campus residence were approved by the Institutional Review Board at the University of Michigan (HUM185966).

Genome Amplification, Sequencing, and Consensus Generation

We sequenced SARS-CoV-2 genomes as described in Valesano et al. [20]. Briefly, we extracted RNA from nasopharyngeal specimens with the PureLink RNA kit and reverse-transcribed RNA with SuperScript IV. We amplified SARS-CoV-2 cDNA in 2 pools using the ARTIC Network v3 primers and protocol. We combined polymerase chain reaction products of each pool in equal volumes for a given sample and purified with AMPure beads. We prepared libraries for sequencing with the NEBNext Ultra II Library Prep Kit. We quantified pooled libraries with a Qubit and sequenced on an Illumina MiSeq (v2 chemistry, 2 × 250 cycles). We mapped reads to the Wuhan/Hu-1/2019 reference genome (GenBank MN908947.3) with BWA-MEM [21]. We trimmed the amplification primer sequences and determined consensus sequences with iVar 1.2.1 [22], using bases with >50% frequency and placing an N at positions covered by <10 reads. Genomes with ≥29 000 unambiguous bases (>97%) were used in downstream analysis.

Case Definitions and Metadata

We used unique identifiers associated with each sample to obtain a single genome per individual and determine which individuals were students in the fall semester (undergraduate, graduate, or professional). We queried 2 databases hosted by the University Health Service to identify students and housing status. Students with a campus residence hall listed in either database were considered "on campus," and the rest were considered "unknown." We obtained case metrics in students from the university COVID-19 Dashboard (https://campusblueprint.umich.edu/dashboard/). We obtained data on new confirmed cases in Washtenaw County and Region 2S from the Michigan Department of Health and Human Services website (accessed March 4, 2021).

Phylogenetic Analysis and Discrete Trait Reconstruction

We subsampled genomes on GISAID using augur [23], excluding those with <27 000 bases, and aligned with MAFFT. We prioritized genomes with genetic similarity to Michigan sequences as follows: 100 genomes per month in Michigan, 5 genomes per month per division in the United States outside of Michigan, 1 genome per country per month in North America outside of the United States, and 2 genomes per country per month outside of North America. We included all complete genomes sequenced here (n = 1659) and all other genomes from Michigan collected in August–December 2020. This resulted in an alignment with 7174 sequences, including 3318 genomes from Michigan, 2157 genomes from the United States outside of Michigan, and 1699 global genomes.

We masked the 5' and 3' ends of the alignment along with other sites commonly affected by sequencing errors and homoplasies. We inferred a maximum likelihood phylogeny with IQ-TREE with a generalized time reversible model [24] and used TreeTime to generate a time-scaled phylogeny rooted on Wuhan/Hu-1/2019 with a clock rate of 0.0008 with a standard deviation of 0.0004 substitutions/site/day [25]. We used TempEst to fit and plot a root-to-tip regression of divergence over time [26]. We removed genomes that exceeded 3 interquartile ranges from the root-to-tip regression (n = 25, including 2 genomes we generated). We generated a new timescaled tree with this filtered alignment (n = 7149). We used it as the basis for discrete-trait ancestral state reconstruction with TreeTime and BEAST 1.10.4 [27, 28]. We inferred introductions from nonstudent to student populations at nodes with a BEAST probability of >0.9. To determine whether the contextual genomes were biasing our results, we generated a total of 10 random subsamples of the global data using the schema described above and analyzed the data in the same manner (Supplementary Figure 2).

Availability of Data and Materials

Consensus genomes generated for this study are available on GISAID. Accessions for all genomes are listed in Supplementary Table 1. Analysis code is available at https://github.com/lauringlab/SARS2_Fall_2020.

RESULTS

We initiated prospective genomic surveillance in Southeastern Michigan in August 2020 with the goal of capturing SARS-CoV-2 transmission dynamics at the University of Michigan-Ann Arbor and the surrounding community. We obtained and sequenced all available SARS-CoV-2-positive specimens from the Michigan Medicine Clinical Microbiology Laboratory and the University Health Service (UHS) on a daily basis, from August 16, 2020, through November 24, 2020. November 24 corresponded to the end of in-person instruction for the

semester and the peak of the November surge of new cases in Washtenaw County. The Clinical Microbiology Laboratory has performed COVID-19 testing for all inpatient and ambulatory clinical settings associated with Michigan Medicine, a large academic medical center with $\sim\!2.3$ million patient clinic visits annually. Testing of students presenting to UHS was either performed on site or sent to the Clinical Microbiology Laboratory. These 2 specimen sources allowed us to broadly sample the university student population as well as individuals from the community.

University Setting and Epidemic Course

The COVID-19 epidemic in Southeastern Michigan evolved over the Fall 2020 semester. Washtenaw County (population 367 000) is part of Michigan Public Health Preparedness Region 2 South, which includes Monroe and Wayne counties and the city of Detroit. The State of Michigan was under a face mask requirement throughout fall of 2020. In addition, gathering sizes at private residences were restricted through state and county orders to no more than 10 people indoors and no more than 25 outdoors, and restaurant capacity was limited to 50%. From August 16 through November 24, there were 6707 laboratory-confirmed COVID-19 cases in Washtenaw County. COVID-19 cases increased in Washtenaw County over the fall semester (Figure 1A), rising from 130 cases during the week of August 23 (63 daily cases per million) to 1125 cases during the week of November 15 (405 daily cases per million).

The University of Michigan–Ann Arbor, a public university with undergraduate, graduate, and professional students, is centrally located in the city of Ann Arbor. In Fall 2020, the university opened for on-campus student residence and hybrid in-person and remote learning. There were 12 812 students residing in Washtenaw County for Fall term, with 1736 in campus

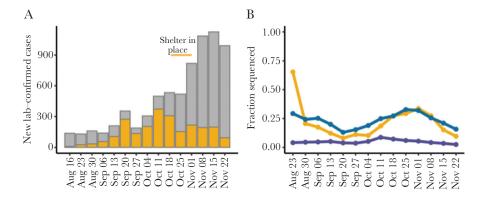


Figure 1. Case curves and sequencing density. A, New lab-confirmed cases of COVID-19 in Washtenaw County, Michigan, from the week of 8/16/2020 through 11/23/2020, displayed by day of symptom onset (as reported by MDHHS). New cases per week are shown on the y-axis and time in weeks on the x-axis. The fraction of new lab-confirmed cases in University of Michigan students is shown in yellow. The Washtenaw County "shelter-in-place" order for undergraduates is indicated (10/20/2020 through 11/03/2020). B, Sampling density is displayed as the fraction of new lab-confirmed COVID-19 cases with complete genome sequences (y-axis) per week during the fall term (x-axis). The fraction of student cases sequenced is shown in yellow, all Washtenaw County cases in blue, and all cases in Region 2S in violet (includes Washtenaw, Wayne, and Monroe counties). Abbreviations: COVID-19, coronavirus disease 2019; MDHHS, Michigan Department of Health and Human Services.

housing. Campus mitigation measures included August prearrival testing of students, daily symptom checks and reporting through a phone app, asymptomatic testing of a subset of students, symptomatic testing, isolation of cases and quarantine of close contacts, reduced residence hall and classroom capacity, and reduced occupancy in university studios and laboratories. In-person instruction was held from August 31 through November 24.

Out of 72 798 tests performed from the week of August 16 through the week of November 22, there were 2374 COVID-19 cases in students, 1064 (44.8%) of which had testing performed at Michigan Medicine laboratories. There was no syndromic case definition for obtaining COVID-19 testing at that time, and testing was available for both symptomatic and asymptomatic cases depending on clinical circumstances. Cases in University of Michigan-Ann Arbor students constituted the majority of cases in Washtenaw County by the end of September 2020 (Figure 1A). In response to this spike, additional mitigation efforts were implemented, including stay-in-place orders targeted to specific residence halls, mass testing events, and a broad stay-in-place order for all undergraduates from October 20 to November 3. The fraction of campus-associated cases in the county declined after mid-October 2020 during a wave of new infections in Washtenaw County in November 2020.

Genomic Surveillance in Southeastern Michigan

We assembled complete SARS-CoV-2 genomes from 1659 individuals. This represents a median of 24% of the confirmed cases per week from Washtenaw County and a median of 4.5% of cases per week from Michigan Region 2 South (Figure 1B). We sequenced 468 complete genomes from University of Michigan–Ann Arbor students, representing 20% of cases from August 16 through November 24 (Supplementary Figure 1). We were able to determine on-campus residences for 131 of these students; the on- or off-campus residence of the remainder is unknown. The genomes presented here consisted of several different viral clades, mostly Nextstrain clades 20A, 20C, and 20G (Supplementary Figure 1).

We used phylogenetic analysis to characterize the influx of viruses into the student population. We generated a time-calibrated maximum likelihood phylogenetic tree with our sequenced genomes and additional contextual genomes (see the "Methods" section; Supplemental Figure 2). To optimize the inference of transmission lineages in Southeastern Michigan, we included all available genomes from Michigan on GISAID that were collected from July to December 2020. We inferred traits of ancestral nodes on this time-calibrated phylogenetic tree using a binary discrete trait model of student vs nonstudent, as has been performed in related studies [27, 29–31]. Genomes from students that shared the same ancestral "student" node were considered part of the same introduced transmission lineage. Genomes from students that were not preceded by a

"student" node were considered singleton introductions. We verified that our results were not substantially biased by contextual genomes using the same analysis on 10 random subsamples of non-Michigan genomes [29].

Using this approach, we inferred 203 distinct transmission introductions into students (Supplementary Figure 2). There were 2 large transmission lineages in students, which we denote as Cluster A (n = 115 students) and Cluster B (n = 73 students). These were the predominant source of cases in students during the middle of the semester, representing >50% of genomes from students from the week of September 20 through the week of October 18. The rest of the introductions were phylogenetic singletons (n = 171) or small clusters of 2–8 students (n = 30 introductions). Small transmission lineages (2–8 students) were often short in duration, lasting a median (IQR [interquartile range]) of 3.5 (1–8) days. The frequency of singletons increased during the latter half of the semester (Figure 2B), indicating new introductions into the student population rather than spread from older transmission clusters, such as Clusters A and B. These data suggest a shifting epidemiology in the student population, characterized by 2 dominant lineages during the early and middle portions of the semester followed by many small introductions later in the semester as community incidence increased.

We further investigated the 2 largest inferred lineages, which indicate sustained local transmission of SARS-CoV-2 within the student population. Genomes from Cluster A were part of Nextstrain clade 20B and Pango lineage B.1.1.304 (TMRCA, August 26⁵ 95% CI, August 3-September 26). Genomes from Cluster B were part of Nextstrain clade 20C and Pango lineage B.1.593 (TMRCA, July 29; 95% CI, July 3-August 28). We detected cases from Cluster A from October 8 through November 14 and Cluster B from September 11 through October 23 (Figure 2C). These clusters were genetically distinct from other lineages in our data and were identified in all 10 independent subsamples (Supplementary Figure 2). Lineages B.1.1.304 and B.1.593 are rare in the GISAID database (0.07% and 0.02%, respectively, out of 532 715 genomes from the United States as of June 7, 2021). Outside of the genomes presented here, B.1.1.304 has been detected in Michigan only 3 times (238 in the United States), and B.1.593 has not been detected elsewhere in Michigan (only twice in the United States). The rarity of these lineages suggests that these outbreaks among students did not spark significant transmission in Southeastern Michigan or in other parts of the country.

We examined the on-campus residence locations for students in these transmission lineages. There was no obvious association with a single residence hall in any transmission lineage in students (Figure 2C). Of the 131 individuals with known on-campus residences, 31% were singleton introductions (n = 40) and 56% (n = 73) were part of Cluster A or B. Individuals from many different residence halls dispersed across the campus were

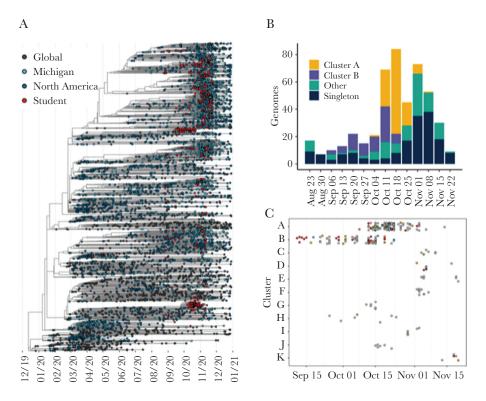


Figure 2. Introductions into the student population. A, Time-calibrated maximum likelihood phylogenetic tree of 7149 SARS-CoV-2 genomes, including 1657 genomes sequenced for this study. Month is shown on the x-axis. Genomes from students are shown in red, other genomes from Michigan in light blue, other genomes from North America in dark blue, and global genomes in gray. B, Counts of genomes from students (y-axis) per week during the fall term (x-axis) by inferred transmission lineage group. Singleton introductions are shown in dark blue, genomes from Cluster A in yellow, genomes from Cluster B in violet, and genomes from smaller clusters (2−8 individuals) in teal. C, Inferred transmission lineages in students, with genomes from each lineage or cluster (y-axis) shown as points with time on the x-axis. Only inferred transmission lineages with ≥5 students are shown. Point colors reflect on-campus residences, if known (unknown residence is light gray). Abbreviation: SARS-CoV-2, severe acute respiratory syndrome coronavirus 2.

present in Clusters A and B, including students from 9 and 7 residences, respectively. Of the 188 students in Clusters A and B, 61% (n = 115) did not have an identifiable on-campus residence and likely lived off-campus (Figure 2C, light gray points). Besides Clusters A and B, there was only 1 other student lineage that had >1 individual from the same campus residence (2 of 3 students in the lineage were from the same residence). The first individuals detected in Cluster B resided in the same residence hall (Figure 2C), potentially reflecting transmission in a congregate setting. However, without detailed contact tracing to complement the genomic data, it is unclear whether clusters A and B originated in specific residence halls and then spread further among students, or whether the clusters originated in off-campus gatherings. Overall, these data demonstrate local transmission among the student population with intermixing among students from multiple on-campus residences and students residing off-campus.

Limited Spillover From Student Clusters Into the Broader Community

Clusters A and B, defined by Pango lineages B.1.1.304 and B.1.593, waned by mid-November, when new COVID-19 cases increased in Washtenaw County (Figure 3), which suggested that these large, student-dominated lineages did not

significantly contribute to the rise in community cases. To further examine the extent of COVID-19 spread from students into the broader Southeastern Michigan community, we used the same ancestral trait reconstruction method as above (see the "Methods" section). There were very few nonstudents in

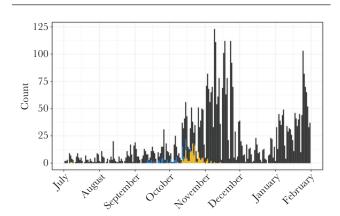


Figure 3. Frequency of lineages B.1.1.304 and B.1.593 in Michigan. All genomes from Michigan available on GISAID from July 2020 to January 2021 are shown. Count of genomes is shown on the y-axis with time on the x-axis (binwidth = 1 day). Genomes from lineage B.1.593 are shown in blue, genomes from B.1.1.304 in yellow, and genomes from all other lineages are shown in gray.

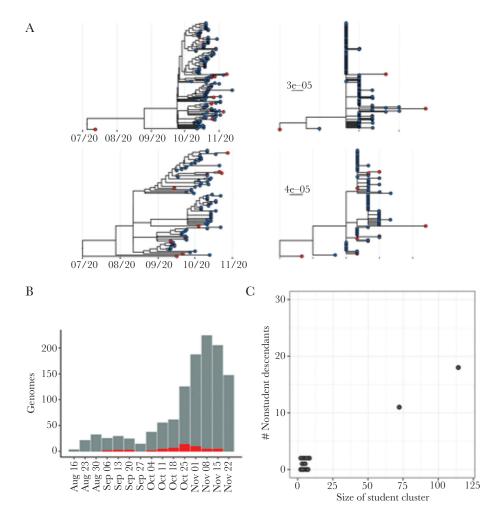


Figure 4. Spillover from student-associated clusters. A, Maximum likelihood phylogenetic trees of student Cluster A (top) and Cluster B (bottom). Time-calibrated trees are displayed on the left and divergence trees on the right. Tip colors reflect genomes from students (blue) and nonstudents (red). B, Bar plot of the number of nonstudent genomes sequenced (y-axis, n = 1191) per week over the fall term (x-axis). Genomes that are derived from inferred "student" nodes are shown in red, and genomes not derived from "student" nodes are shown in gray. C, For each nonsingleton transmission lineage in students, the number of nonstudent descendants is shown (y-axis) by the number of students in the cluster (x-axis).

the community with genomes that descended from studentassociated lineages (n = 53), and most were nested within either Cluster A (n = 18) or B (n = 11) (Figure 4A). A total of 24 nonstudents descended from the rest of the inferred transmission lineages (2-8 students each). Out of the 1191 genomes we sequenced from nonstudents, 96% (n = 1138) were not genetic descendants of detected clusters in students (Figure 4B). Larger clusters of students had greater numbers of nonstudent descendants (Figure 4C). The median age of nonstudents descending from Clusters A and B (IQR) was 47 (20-61) years. We do not have epidemiologic information on the association of these individuals with students, so it is not possible to determine the circumstances of transmission. It is also possible that student status was misclassified for these individuals or that some of these individuals were campus faculty or staff and therefore had differential exposure compared with the broader community.

DISCUSSION

We conducted a prospective study of SARS-CoV-2 genomic surveillance focused on a large public university and the surrounding community. A major strength of our study is high-density sampling of the student population and the surrounding community by sequencing all available specimens from 2 major testing laboratories. These data illustrate the rapid transmission of SARS-CoV-2 within a large public university population with remarkably little spread into the community.

Our analysis demonstrates that the COVID-19 epidemic among students at the University of Michigan–Ann Arbor was not derived from a single introduction. Most student cases in early Fall were derived from 1 of 2 dominant viral lineages, which coexisted for several weeks and circulated throughout several on-campus residences. We think it is unlikely that these 2 Pango lineages (B.1.1.304 and B.1.593) have

enhanced transmissibility or other notable intrinsic properties. They did not disseminate widely and do not exhibit the same mutations as other highly transmissible variants. The influx of new singleton introductions late in the semester may have been driven by "outside-in" transmission from elsewhere in the community as incidence spiked in Washtenaw County. This emphasizes the importance of reducing overall county and regional COVID-19 incidences, in addition to preventing outbreaks at IHE.

A key finding was that very few genomes from nonstudents were genetically linked to student clusters. Notably, lineages B.1.1.304 and B.1.593 have not persisted in Michigan since this time, providing additional support that these outbreaks did not spur widespread transmission in the community. We did not sequence viruses from every infection and therefore cannot exclude that other spillover events may have occurred. Nevertheless, given our notably high depth of sampling, it is unlikely that large outbreaks among students were the source of most COVID-19 infections in the community. If this was the case, we would expect to see vastly different patterns, particularly a higher frequency of B.1.1.304 and B.1.593 in nonstudents during November. Although there are differences in context, setting, and mitigation measures here compared with other IHE, this study suggests that previous findings with limited community surveillance may generalize more broadly [4, 6, 18].

There are other important limitations. First, we were not able to access specimens from commercial testing sites, and the number of detected transmission introductions is certainly an underestimate. Next, our study is not an epidemiologic investigation with contact tracing and individual behavioral information. It is difficult to reliably assess the effectiveness of any single mitigation measure from these data alone. Thorough contact tracing investigations of IHE-associated outbreaks with dense genomic surveillance across populations may be able to resolve these questions in greater detail. It is also possible that these dynamics could have played out differently with earlier emergence of a highly transmissible variant, such as B.1.1.7 [27, 32]. This work will be a valuable point of comparison for future studies examining the effects of more transmissible variants and vaccination on COVID-19 spread within IHE.

Our phylogenetic analysis of well-sampled genomic surveillance data provides insight into the spread of SARS-CoV-2 at a large public university in the United States. The small number of dominant lineages that circulated in the early and middle portions of the semester did not significantly contribute to the rise in county-level cases in November 2020. We emphasize that even rare transmission events can disproportionally impact vulnerable populations [17]. Additionally, SARS-CoV-2 infection can have severe clinical manifestations even in populations with generally lower risk [33, 34]. Therefore, it is critical that every effort be made to prevent and mitigate IHE-associated outbreaks in conjunction with measures at broader geographic levels.

Supplementary Data

Supplementary materials are available at *Open Forum Infectious Diseases* online. Consisting of data provided by the authors to benefit the reader, the posted materials are not copyedited and are the sole responsibility of the authors, so questions or comments should be addressed to the corresponding author.

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